

COMP90049 Knowledge Technologies

Introduction: Machine learning and Data Mining (Lecture Set 1)

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The course content will be as follows:

- ☐ Introduction to Machine Learning/ Data Mining
- motivating when you have to use such techniques
- Basic introduction to probability theory
- Basic machine leaning techniques
 - NB
 - Decision Trees
 - SVMs



- □ Pattern mining and how they can be used
 - Frequent pattern mining
 - Association rule
 - Feature selection
- □ Clustering techniques
- □ Evaluation measures of various machine learning techniques: E.g. Accuracy,
 - F-measure, AUC and Cross validation methods.



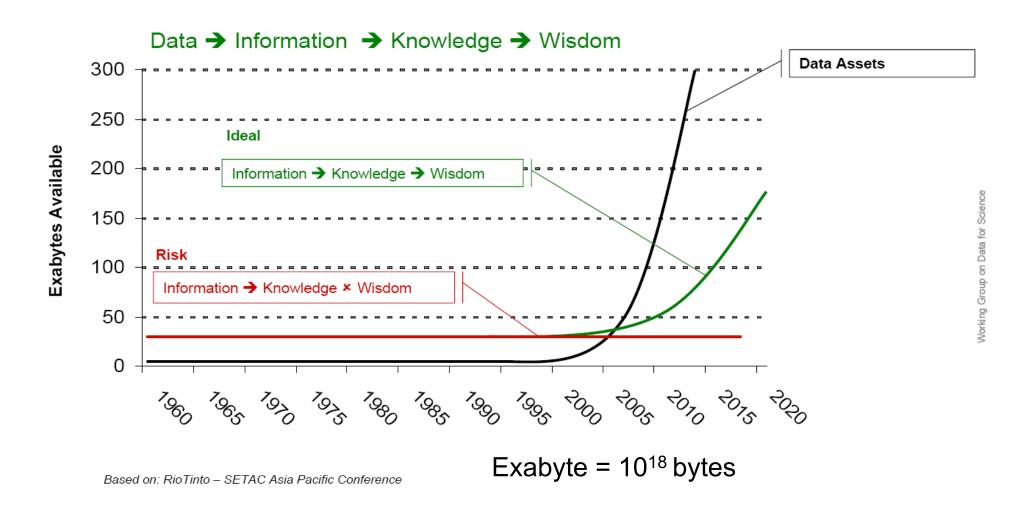


FROM DATA TO WISDOM:

Pathways to Successful Data Management for Australian Science
Working Group on Data for Science
Report to PMSEIC
December 2006

Reference:

www.dest.gov.au/sectors/science_innovation/publications_resources/profiles/documents/Data_for_Science_pdf.htm





Need for data collection and Knowledge acquisition





















Diversity of data sources





Tackling the challenge of knowledge management and discovery at a massive scale

- Database modelling and integration has long been a focus of Information Technology research and development. Classic example being the application of RDBMs for commercial applications.
- A major and accelerating trend is the focus of data integration from business and enterprise applications to scientific and personal applications.



Tackling the challenge of knowledge management and discovery at a massive scale

- Exponential growth of data with the spread of the Internet, Web and the multitudes of automatic data generation and collection devices. Internet of Things (IoT).
- This trend is expected to continue in the foreseeable future.



Life sciences research generating extremely rich and complex datasets

- Terabytes/day (and growing) MRI (magnetic resonance Imaging), next generation sequencing, metabolomics, large public projects linking research and clinical data
- Life Technologies announced in 2012 that they can already scan the full genome for \$1000, and another company called Geniachip claims to go beyond this and deliver the same results for just \$100 cheaper than MRI!



Life sciences research generating extremely rich and complex datasets

- Very high dimensionality, several hundreds of millions of data points
- E.g., world population 7.5 billion and each person has Genome information of 3 billion DNA sequences
- Different formats and structures: spatial/temporal/sequence/graph/streams/medical images/video/audio/...

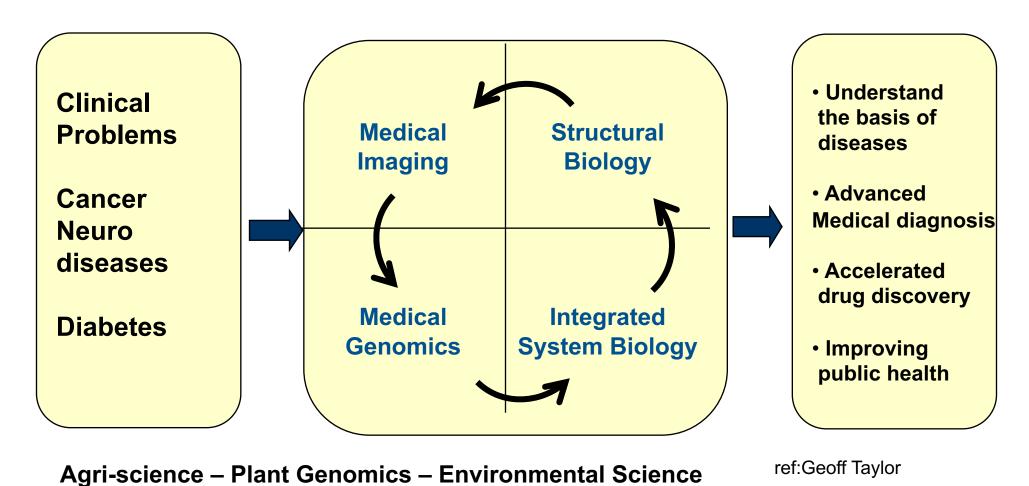


Importance of Problem

- Current computational methods cannot handle magnitude and dimensionality of the data
- Decision makers and Scientists need techniques to help form hypotheses and make evidence based decisions



Example





Research issues

- Data management
 - Issues: Scale, Data Quality: incomplete, inconsistent, multi source, ...
 - Provide answers to user queries.
- Machine Learning and Data mining
 - Issues: Scale, dimensionality, data modalities (relational, graphs, text, in addition to problems faced by data management systems, etc.)
 - Discover models that fit the data, explain underlying processes that generated the in information.



Some Research issues

- Databases and information retrieval
 - Issues: Scale, Query Languages,
 - Data modalities (relational, graphs, text, etc.)
- Distributed computing: grid, peer-to-peer and cloud computing models
 - Issues: parallelization (algorithmic aspects), distribution,
 - load balancing, robustness, recovery, ...
 - Cloud computing is already making inroads and is going to dominate in the immediate future as computational utility.
- Security, Privacy and Trust
 - Issues: scale, management, policy.





Applications

Business Applications (banks, supermarkets, Govt.)



Scientific Applications

(Biomedical, Water Resources Management)

Data Analysis & Actionable Knowledge Discovery

Data Management

(managing data volume, complexity, quality and semantics)

Infrastructure: E.g. Cloud Computing and Services

Big pictures of Knowledge Technologies



Multi Disciplinary Scope

Cross disciplinary across many fields:

Engineering, Medicine, Commerce, etc.



Grand Challenges in Computational Biology Joint BSC - IRB Barcelona Conference Barcelona, 2-4 June 2008

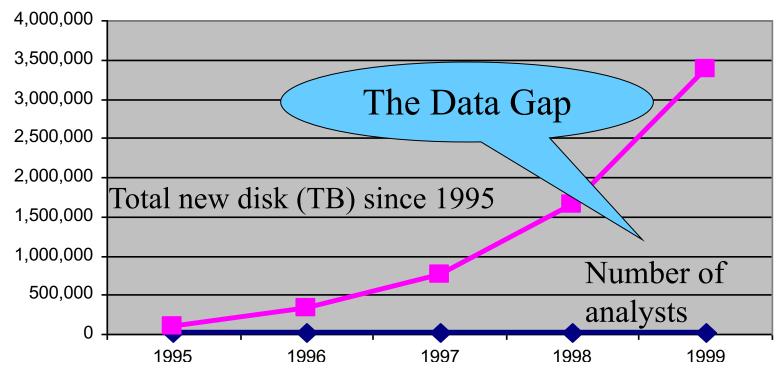
"There are very few cases of human activities growing faster than Moore's law (doubling number of transistors every 18 months), and biology is one of them. The exponential growth of sequence and structural databases, and the discovery of the complexity of most biomolecular interactions are giving rise to computational challenges ..."

"These problem can be tackled by employing computational knowledge discovery tools."



Mining Large Data Sets - Motivation

- There is often information "hidden" in the data that is not readily evident
- Human analysts may take weeks to discover useful information
- Much of the data is never analyzed at all



From: R. Grossman, C. Kamath, V. Kumar, "Data Mining for Scientific and Engineering Applications"



Application areas of Machine learning and Data Mining:

- Bioinformatics
- Commerce
- Health
- Engineering
- Information retrieval (IR)
- Modelling complex systems
- Decision support systems
- Spam detection
- Language Technologies
- Almost all fields of science, engineering, life sciences, social sciences, ...



Summary

- Exponential growth in collection and availability of data
- Society enforces decision makers to make decisions based on evidence
- Need effective tools to manage and process information
- •Trust worthiness of information will become a major issue. Can computational methods ever determine trust?
- Data mining/Machine learning tools are required for processing a variety of complex data for aiding in decision making